# TASK-2: Prediction using Unsupervised ML - Spark Foundation

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# **Import Dependencies**

```
In [1]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

#### In [2]:

```
data = pd.read_csv('iris.csv')
```

## In [3]:

```
data.head()
```

#### Out[3]:

	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

#### In [4]:

```
data['Species'].value_counts()
```

#### Out[4]:

Iris-virginica 50
Iris-versicolor 50
Iris-setosa 50
Name: Species, dtype: int64

#### In [5]:

```
features = data.iloc[: , :-1].values
```

# K-Means

#### In [6]:

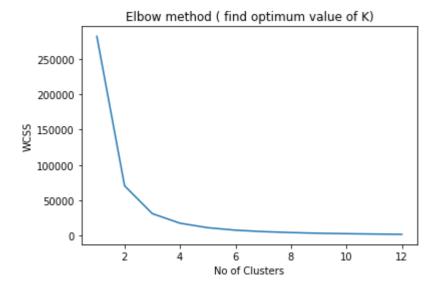
```
from sklearn.cluster import KMeans
```

#### In [7]:

```
wcss = [] ## Within-Cluster Sum of Square
for i in range(1,13):
    kmeans = KMeans(init='k-means++' , n_clusters = i , random_state = 0)
    kmeans.fit(features)
    wcss.append(kmeans.inertia_)
```

#### In [8]:

```
plt.plot(range(1,13) , wcss)
plt.title('Elbow method ( find optimum value of K)')
plt.xlabel('No of Clusters')
plt.ylabel('WCSS')
plt.show()
```



## In [9]:

```
## Optimum Wcss point where cluster == 3.
## so we take n_clusters = 3 and make the model
```

# In [10]:

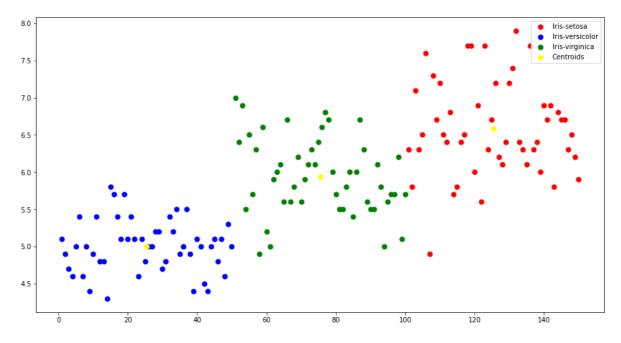
```
kmeans = KMeans(init='k-means++' , n_clusters = 3 ,random_state = 0)
kmeans.fit(features)
prediction = kmeans.predict(features)
```

# **Final Clusters**

#### In [12]:

# Out[12]:

<matplotlib.legend.Legend at 0x11fa9ced3a0>



#### In [ ]: